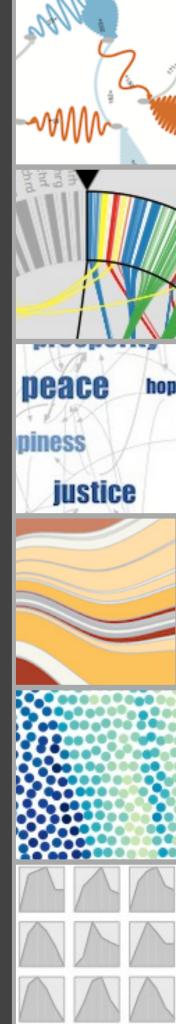
cs6964 | January 19 2012

DATA

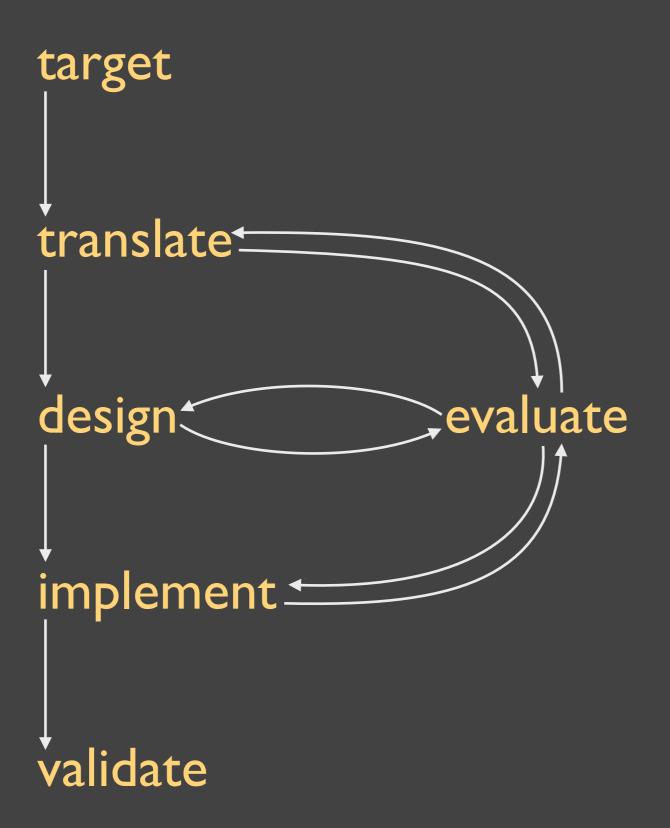
Miriah Meyer
University of Utah

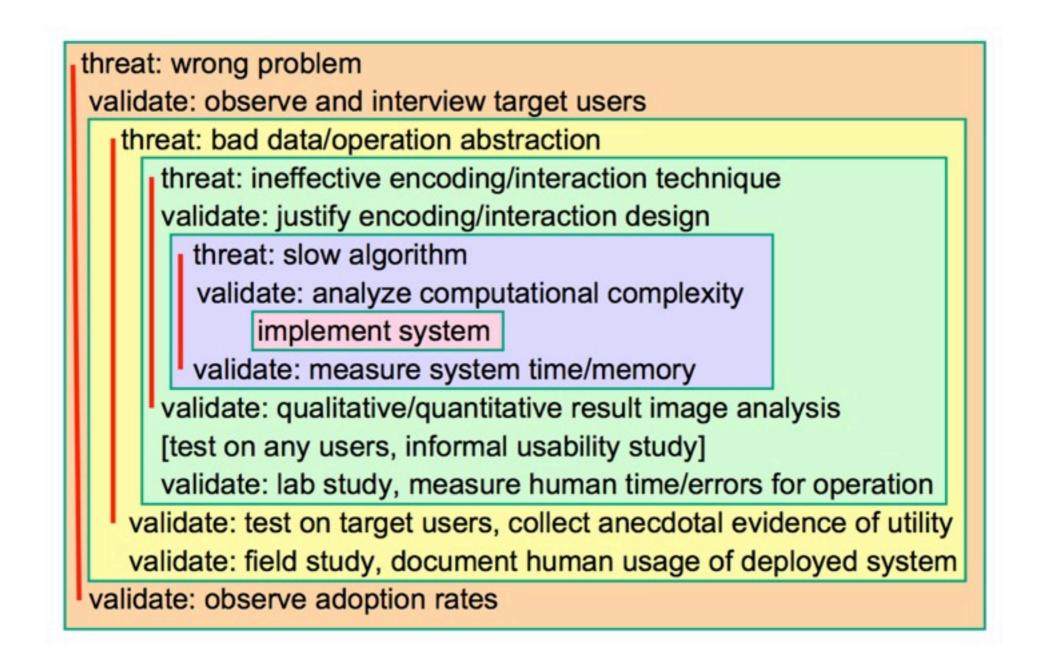
slide acknowledgements:

Tamara Munzner, University of British Columbia Hanspeter Pfister, Harvard University



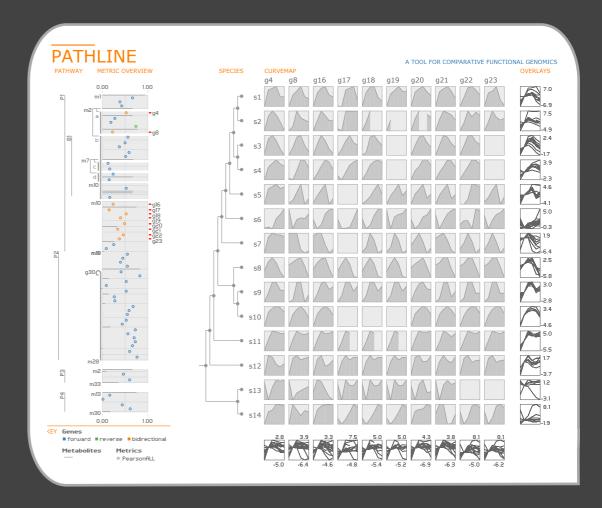
LASTTIME





avoid validation mismatch

- -cannot validate encoding with system timings
- -cannot validate abstraction with lab studies



Pathline

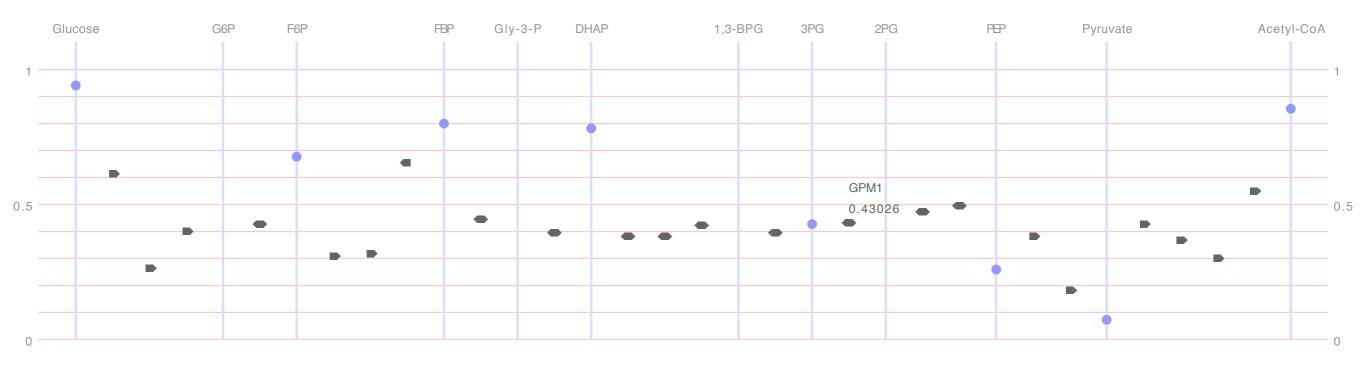
A Tool for Comparative Functional Genomics Data

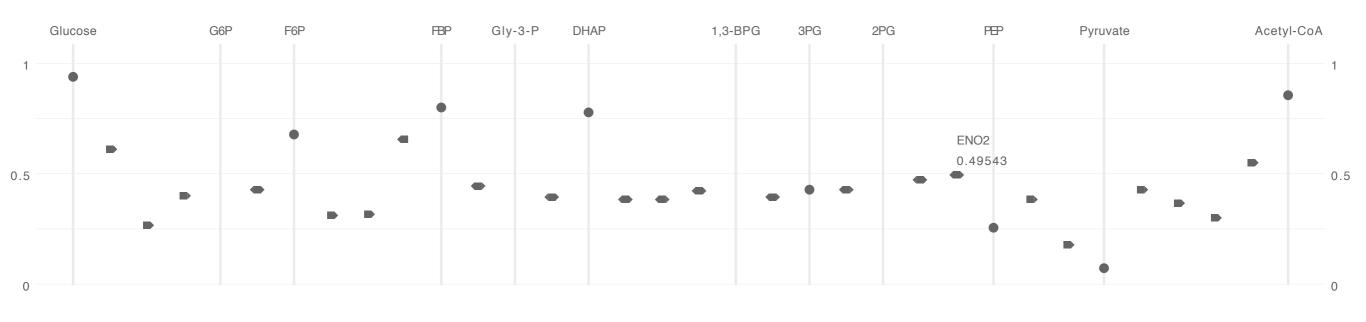
joint work with:

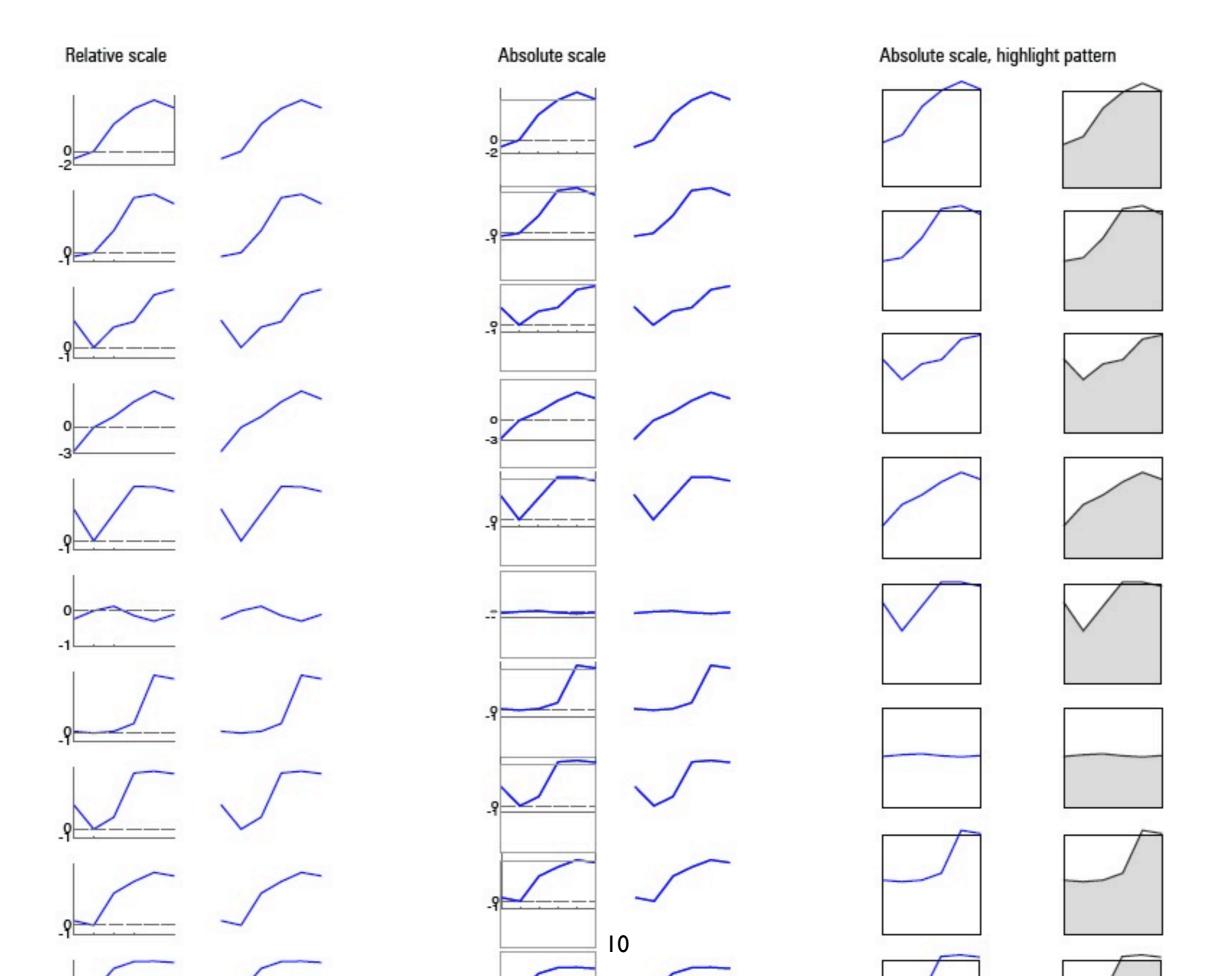
Bang Wong, Mark Styczynski, Tamara Munzner, Hanspeter Pfister

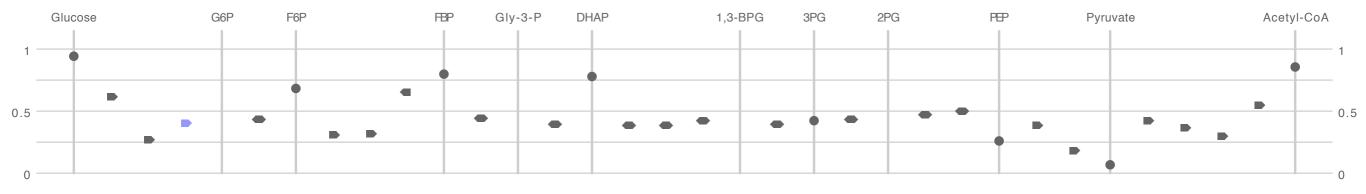
LESSONS LEARNED

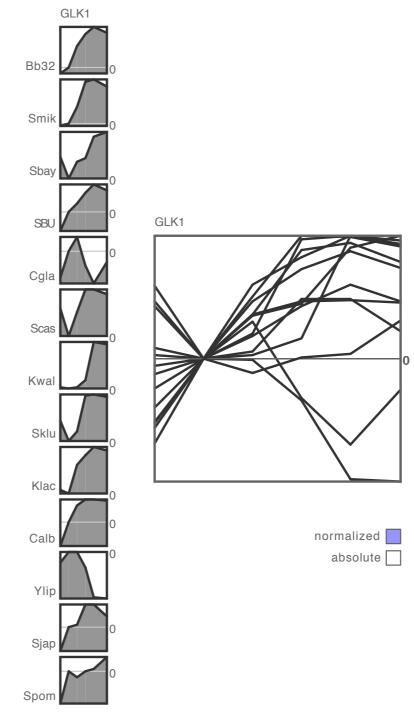
- -process supports efficient development
- -collaborators' time commitment is front loaded
- -rapid prototyping is essential

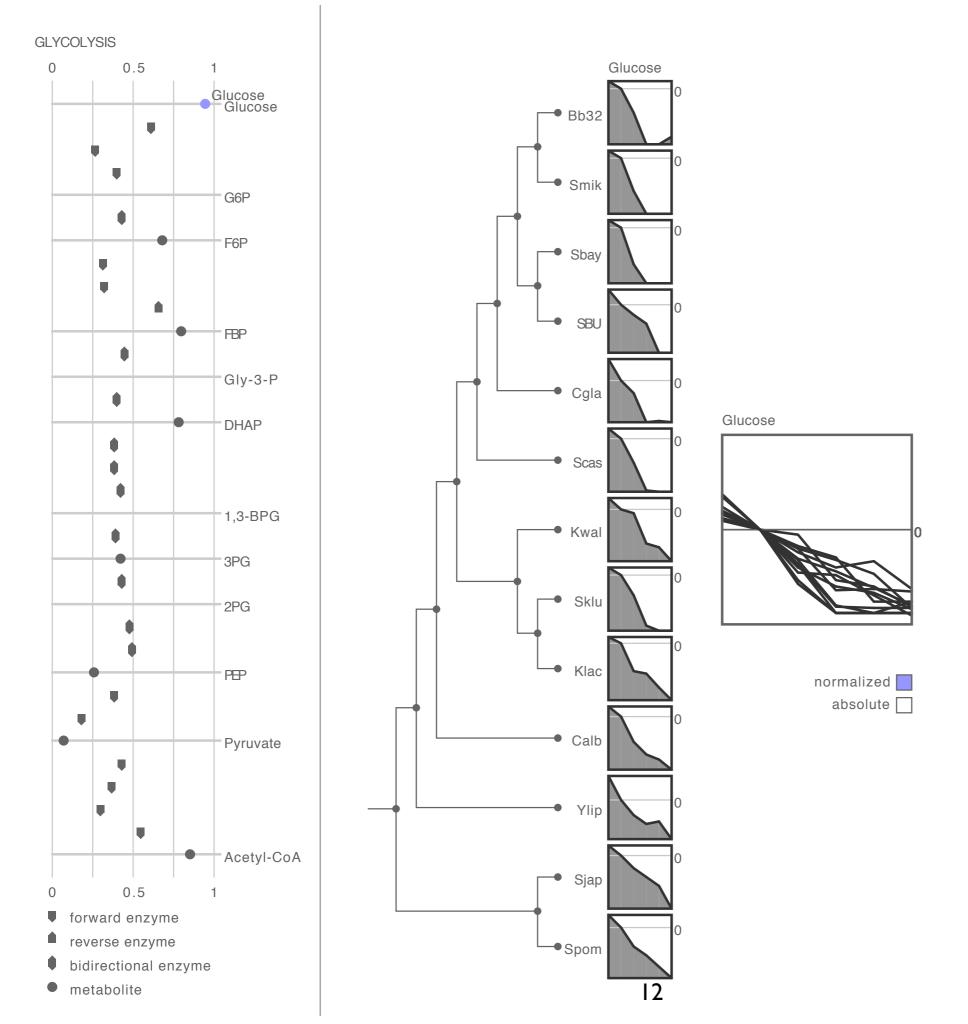


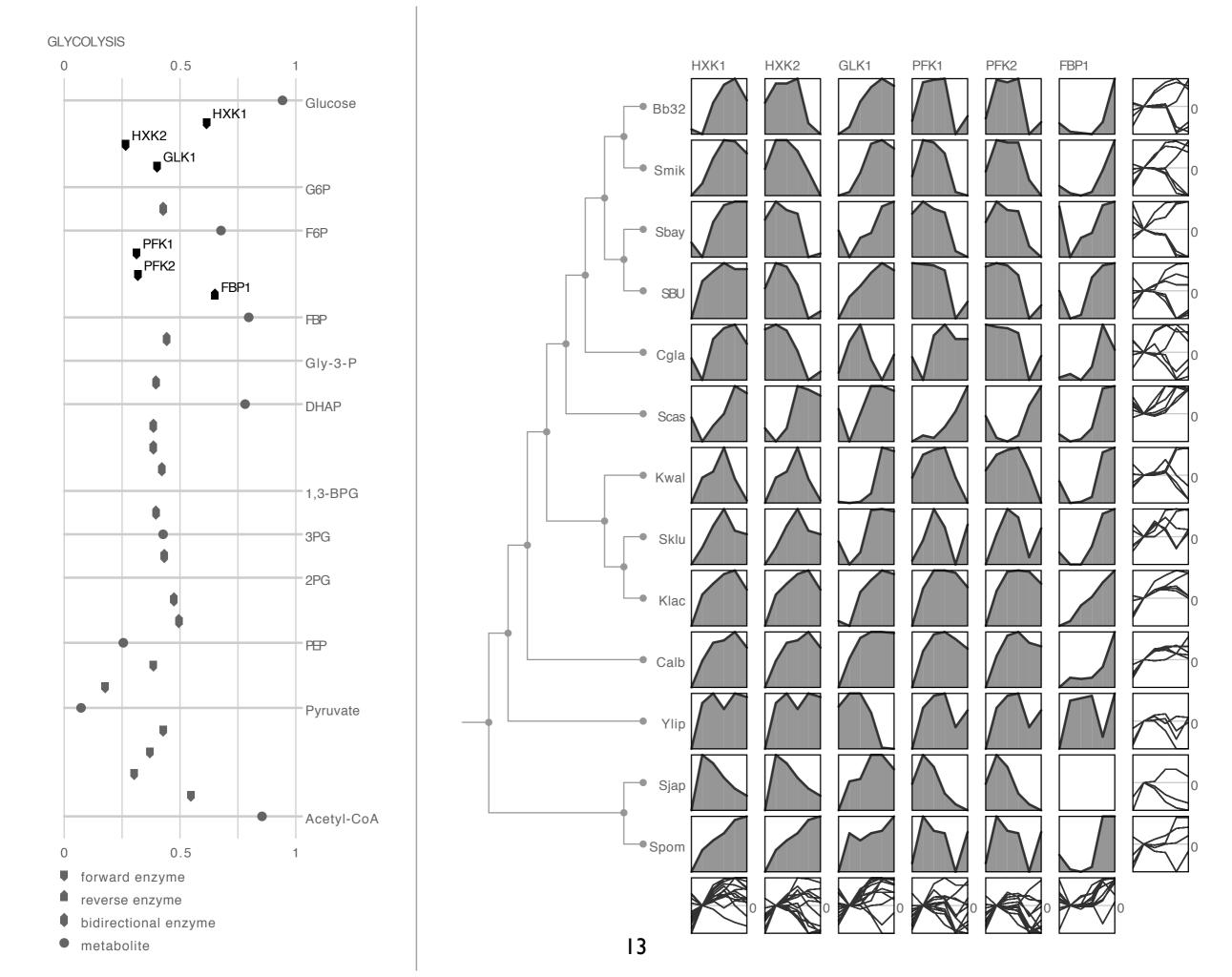


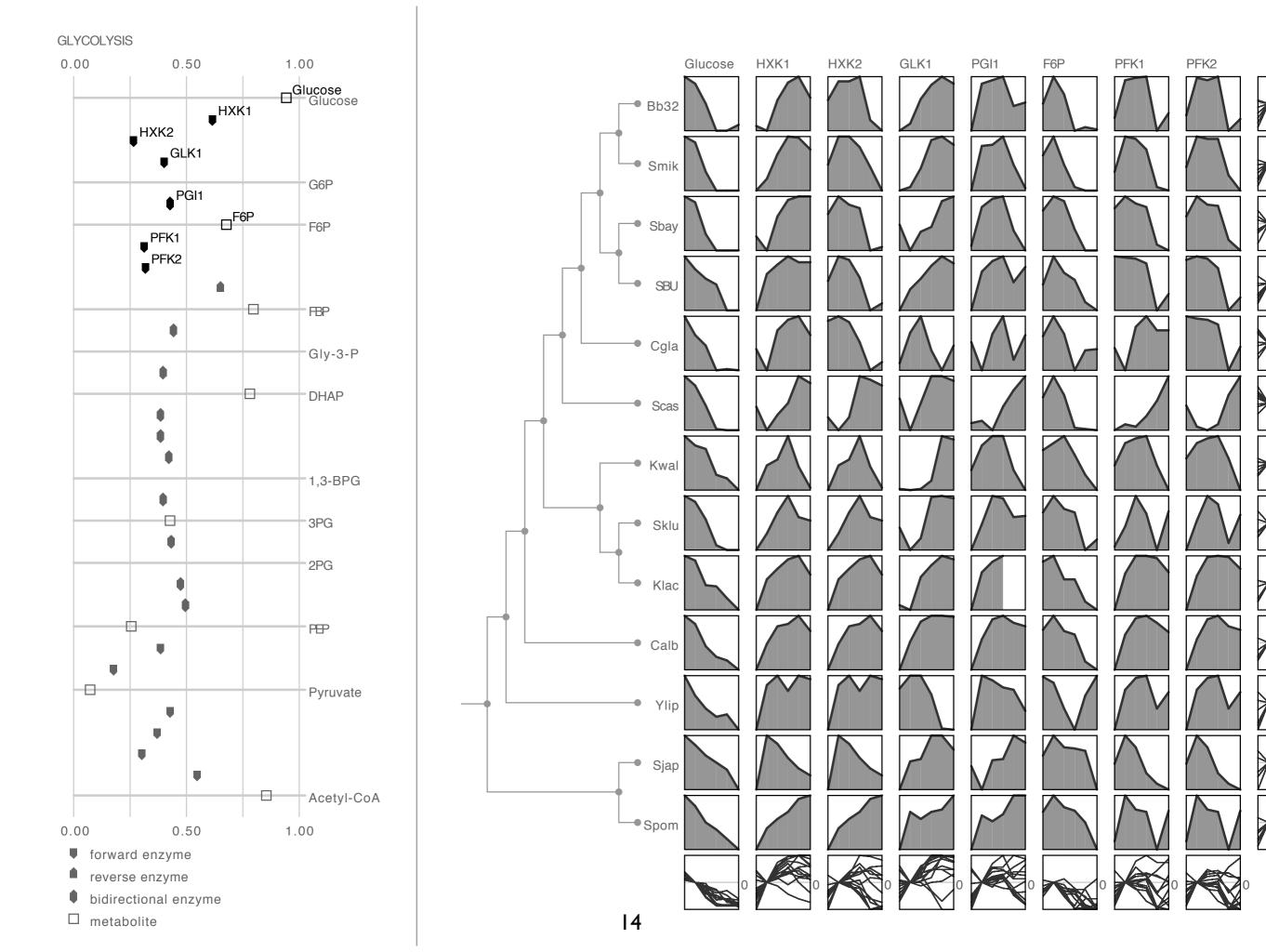


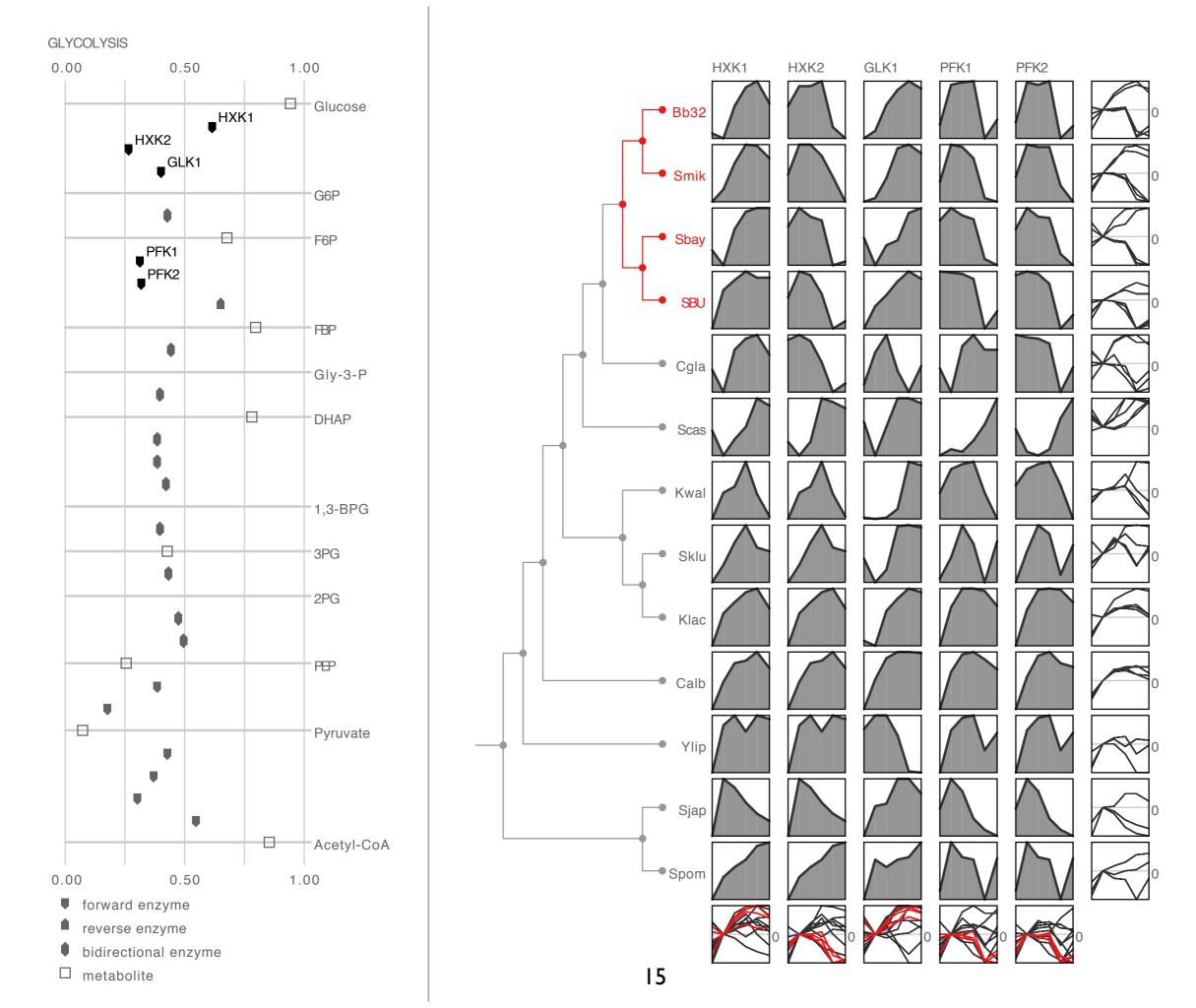


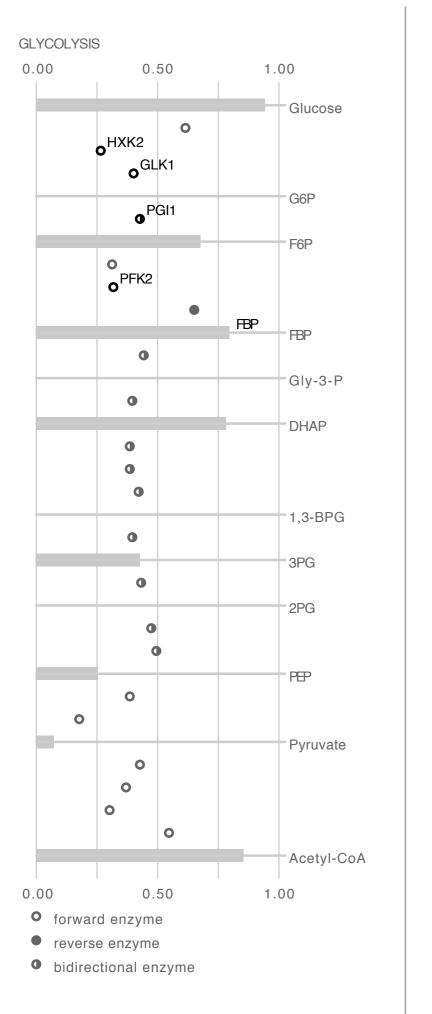


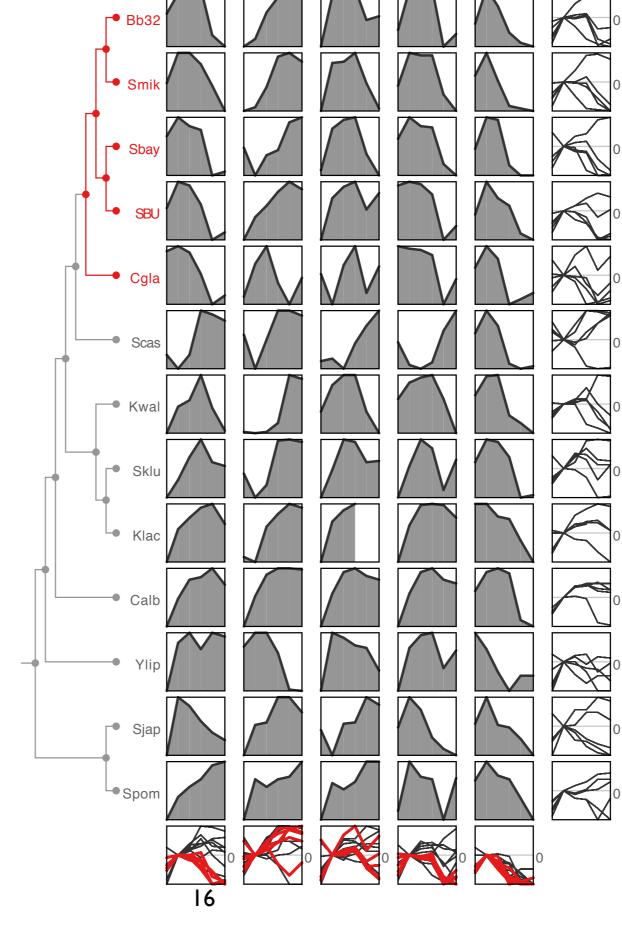












HXK2

GLK1

PGI1

PFK2

FBP

LESSONS LEARNED

- -process supports efficient development
- -collaborators' time commitment is front loaded
- -rapid prototyping is essential
- -put off coding as long as possible

contributions

- -Pathline
 - -multiple genes, time points, species, and pathways
- -linearized pathway representation
- -curvemap
- -tool deployment
 - -open source
 - -used daily by several collaborators

- -visualization design process
- -types of research contributions

5.- PAPER TYPES

A Visweek paper typically falls into one of five categories: technique, system, design study, evaluation, or model. We briefly discuss these categories below. Although your main paper type has to be specified during the paper submission process, papers can include elements of more than one of these categories. Please see "Process and Pitfalls in Writing Information Visualization Research Papers" by Tamara Munzner for more detailed discussion on how to write a successful Visweek paper.

Technique papers introduce novel techniques or algorithms that have not previously appeared in the literature, or that significantly extend known techniques or algorithms, for example by scaling to datasets of much larger size than before or by generalizing a technique to a larger class of uses. The technique or algorithm description provided in the paper should be complete enough that a competent graduate student in visualization could implement the work, and the authors should create a prototype implementation of the methods. Relevant previous work must be referenced, and the advantage of the new methods over it should be clearly demonstrated. There should be a discussion of the tasks and datasets for which this new method is appropriate, and its limitations. Evaluation through informal or formal user studies, or other methods, will often serve to strengthen the paper, but are not mandatory.

System papers present a blend of algorithms, technical requirements, user requirements, and design that solves a major problem. The system that is described is both novel and important, and has been implemented. The rationale for significant design decisions is provided, and the system is compared to documented, best-of-breed systems already in use. The comparison includes specific discussion of how the described system differs from and is, in some significant respects, superior to those systems. For example, the described system may offer substantial advancements in the performance or usability of visualization systems, or novel capabilities. Every effort should be made to eliminate external factors (such as advances in processor performance, memory sizes or operating system features) that would affect this comparison. For further suggestions, please review "How (and How Not) to Write a Good Systems Paper" by Roy Levin and David Redell, and "Empirical Methods in CS and AI" by Toby

DATA ABSTRACTION

comments on readings?

target translate design implement validate

DATASET TYPES

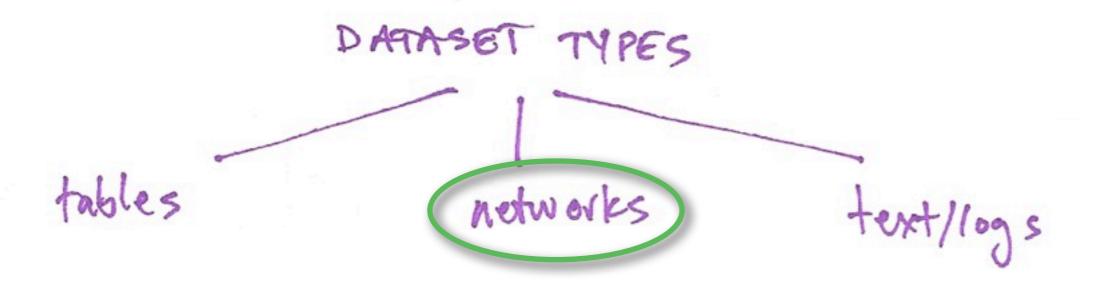
tables

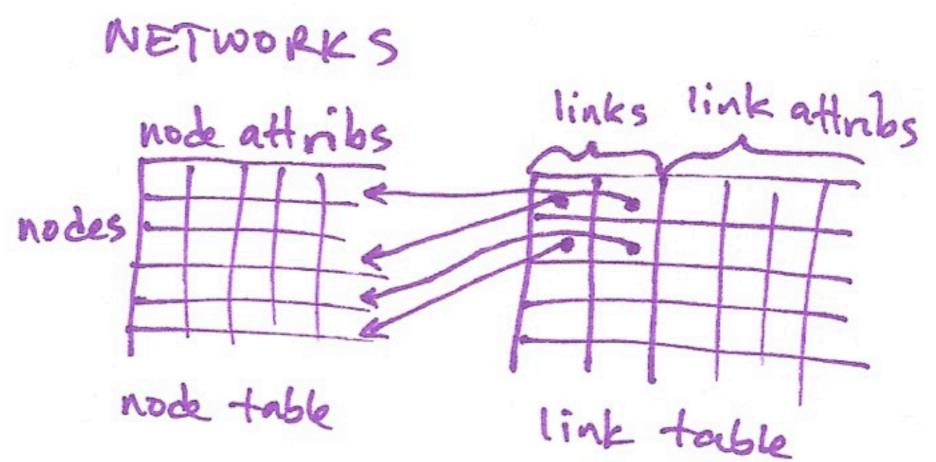
networks

text/logs

TABLES

columns = attributes





graph: another word for networks tree: network without cycles

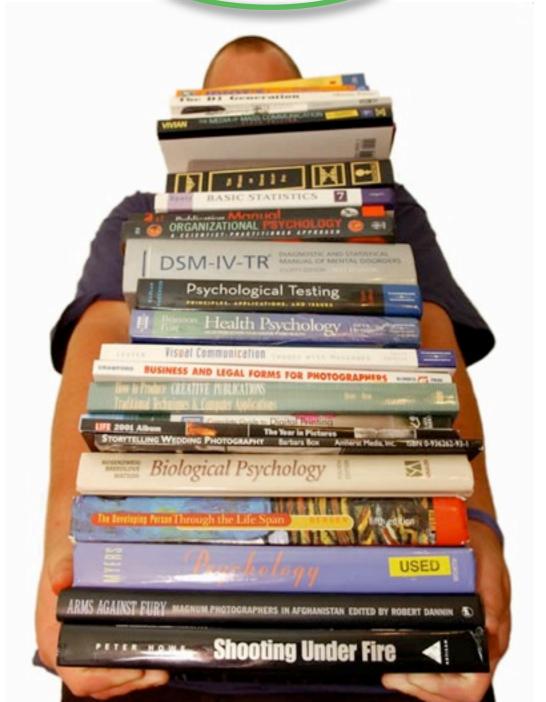
DATASET TYPES

tables

networks

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	1.4.4	A Computer In The Loop	24			
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ATTRIBUTE TUPES mathematical interpretation

Categorical

ordered

no implicit ordering

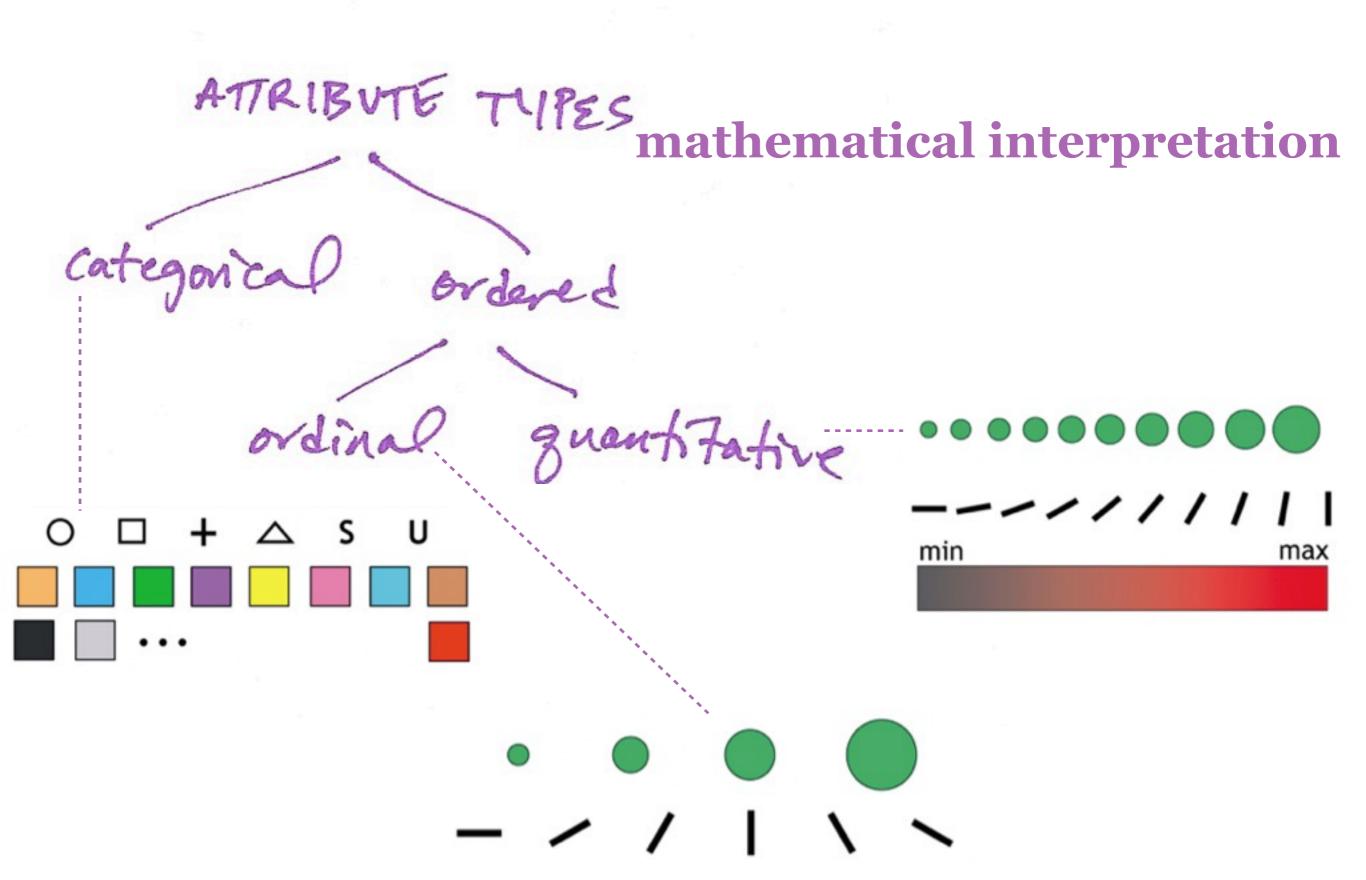
ATTRIBUTE TUPES mathematical interpretation

categorical ordered

ordinal quantitative

meaningful magnitude, can do arithmetic

examples?



ATTRIBUTE TUPES mathematical interpretation diverging

examples?

Α	В	С		S	Т	U
Order ID	Order Date	Order Priority		Product Container	Product Base Margin	
3	10/14/06	5-Low		Large Box	0.8	10/21/06
6	2/21/08	4-Not Speci	fied	Small Pack	0.55	2/22/08
32	7/16/07	2-High		Small Pack	0.79	7/17/07
32	7/16/07	2-High		Jumbo Box	0.72	7/17/07
32	7/16/07	2-High		Medium Box	0.6	7/18/07
32	7/16/07	2-High		Medium Box	0.65	7/18/07
35	10/23/07	4-Not Speci	fied	Wrap Bag	0.52	10/24/07
35	10/23/07	4-Not Speci	fied	Small Box	0.58	10/25/07
36	11/3/07	1-Urgent		Small Box	0.55	11/3/07
65	3/18/07	1-Urgent		Small Pack	0.49	3/19/07
66	1/20/05	5-Low		Wrap Bag	0.56	1/20/05
69	6/4/05	4-Not Speci	fied	Small Pack	0.44	6/6/05
69	6/4/05	4-Not Spec	anar	ntitative	0.6	6/6/05
70	12/18/06	5-Low	quai	ititative	0.59	12/23/06
70	12/18/06	5-Low	ordi	nal	0.82	12/23/06
96	4/17/05	2-High	orar		0.55	4/19/05
97	1/29/06	3-Medium	cates	gorical	0.38	1/30/06
129	11/19/08	5-Low	cates	Sorreur	0.37	11/28/08
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193	8/8/06	1-Urgent		Medium Box	0.57	8/10/06
194	4/5/08	3-Medium		Wrap Bag	0.42	4/7/08

ATTRUB UTE SEMANTICS real-world meaning

ATTRIBUTE SEMANTICS visualization specific

- spatial/nonspatial
- temporal/nontemporal
- independent/dependent *
- continuous/discrete
- Lineusions/measures

- independent/dependent

MULTIDIMENSIONAL TABLES

ind! I ind2 item (dep1, dep2)

ind: index/independent attrib

dep: dependent/value attrib

scalar vector tensor
I depathib 2 depathibs 3 dep

DATASET SEMANTICS visualization specific - spatial/abstract - static/timevarying

DERIVED ATTRIBUTES

- -derived attribute: compute from originals
 - -simple change of type
 - -complex transformation
 - -transformation is abstraction choice



$$h(x) = f(x) - g(x)$$
derived data

DATA vs CONCEPTUAL MODEL

- -data model: mathematical abstraction
 - -set with operations, eg. floats with * / +

- -conceptual model: mental construction
 - -includes semantics, supports reasoning

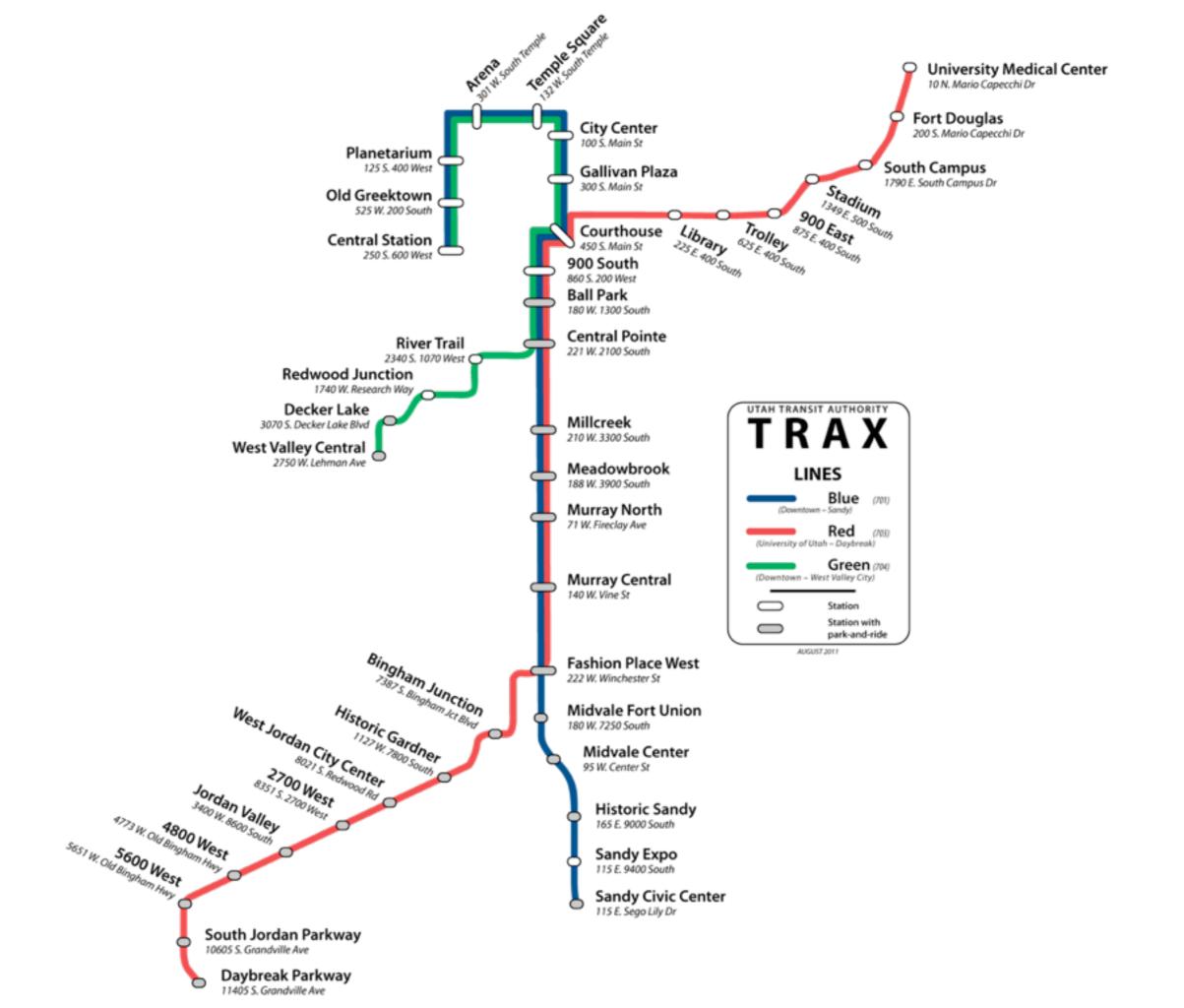
-conceptual model motivates derived data

EXAMPLE

- -from data model . . .
 - -32.52, 54.06, -17.35, ... (floats)
- -using conceptual model . . .
 - -temperature
- -to data type.
 - -continuous to 2 significant figures (Q)
 - -hot, warm, cold (O)
 - -above freezing, below freezing (C)

abstraction exercise ...

ESPRES & HOTT GUATEMALA CASI CIELO Florel, lemon & cocoo GUATEMALA CASI CIELO de-cof Florel, lemon & cocoo SUMATRA Spicy, herbal & earthy	EA 1.50 1.70 1.90 F.6
ESPRESSO	1.45 1.70 1.85
AMERICANO	1.60 1.80 200
CAFÉ LATTE	2.15 2.75 3.20
CAPPUCCINO	2.15 2.75 3.20
CAFÉ MOCHA	2.65 -3.05 3.55
ORGANIC BREAKFAST ORGANIC LONG LIFE GREEN TO MONSOON CHAI CHAI TEA LATTE BLACK TEA LATTE	1.70 1.90 2.10
HOT CHOCOLATE	2:50 2:75 3:00
HOMEMADE SYRUP FL	AVORS .50 each



L₅: Visual Encoding

REQUIRED READING

Visual Encoding Principles

The previous chapter presented a taxonomy of data types; now we discuss how to encode these types with a visual representation. This chapter begins with discussion of human perception as a system for making relative rather than absolute judgements. It continues with the image theory of marks and visual channels, and a discussion of channel types. The chapter presents a ranking of channels according to the data type that they are used to encode, and discusses the concepts of expressiveness and expressiveness. It continues with ways to measure channel effectiveness in terms of accuracy, discriminability, and separability, and ability to provide visual popout. The chapter then discusses the characteristics of each channel, including planar spatial position, color, size, tilt and angle, shape, and stipple. It ends with a discussion of the difficulties of 3D depth coding.

3.1 Relative vs. Absolute Judgements

The human perceptual system is fundamentally based on relative judgements, not absolute ones. Weber's Law states that the amount of difference that we can detect is relative to the context between the two things, not an absolute quantity. ¹ For instance, the amount of length difference we can detect is a percentage of the object's length.

This principle holds true for all sensory modalities. The fact that our

Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design

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Stanford University
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ABSTRACT

Understanding perception is critical to effective visualization design. With its low cost and scalability, crowdsourcing presents an attractive option for evaluating the large design space of visualizations; however, it first requires validation. In this paper, we assess the viability of Amazon's Mechanical Turk as a platform for graphical perception experiments. We replicate previous studies of spatial encoding and luminance contrast and compare our results. We also conduct new experiments on rectangular area perception (as in treemaps or cartograms) and on chart size and gridline spacing. Our results demonstrate that crowdsourced perception experiments are viable and contribute new insights for visualization design. Lastly, we report cost and performance data from our experiments and distill recommendations for the design of crowdsourced studies.

ACM Classification: H5.2 [Information interfaces and presentation]: User Interfaces—Evaluation/Methodology

General Terms: Experimentation, Human Factors.

Keywords: Information visualization, graphical perception,

for ecological validity. Crowdsourced experiments may also substantially reduce both the cost and time to result.

Unfortunately, crowdsourcing introduces new concerns to be addressed before it is credible. Some concerns, such as ecological validity, subject motivation and expertise, apply to any study and have been previously investigated [13, 14, 23]; others, such as display configuration and viewing environment, are specific to visual perception. Crowdsourced perception experiments lack control over many experimental conditions, including display type and size, lighting, and subjects' viewing distance and angle. This loss of control inevitably limits the scope of experiments that reliably can be run. However, there likely remains a substantial subclass of perception experiments for which crowdsourcing can provide reliable empirical data to inform visualization design.

In this work, we investigate if crowdsourced experiments insensitive to environmental context are an adequate tool for graphical perception research. We assess the feasibility of using Amazon's Mechanical Turk to evaluate visualizations and then use these methods to gain new insights into visualization design. We make three primary contributions: